

# Errors in numerical analysis

Numerical analysis - University of Luxembourg - Session 1

Exercises

## Python documentation

- numpy: <https://numpy.org/doc/>
- matplotlib: [https://matplotlib.org/stable/gallery/subplots\\_axes\\_and\\_figures/index.html](https://matplotlib.org/stable/gallery/subplots_axes_and_figures/index.html)
- sympy <https://www.sympy.org/en/index.html>

## Exercises on numerical errors

### Exercise 1. Absolute and relative errors

Step 1: Print the relative and absolute errors of the Stirling approximation

$$n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$

for different values of  $n$ . Explain your observations for large values.

- Implement a function that computes the Stirling approximation `"def Stirling(n): return ..."`, and another one that computes the errors `"def Errors(x, xref): ... return abs err, rel err"`
- For factorials, you can use `"from scipy.special import factorial"`
- For simple plotting, you can use the following concise form
  - `plt.figure(figsize=(7,4))`
  - `plt.plot(N, err0, 'm-o', label = 'Stirling') or plt.loglog(N, err0, 'm-o', label = 'Stirling')`

Step 2: Do again the comparison with more terms of the asymptotic development

$$n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left[1 + \frac{1}{12n} + \frac{1}{288n^2} - \frac{139}{51840n^3} + O\left(\frac{1}{n^4}\right)\right]$$

- You can implement more functions `def Stirling2(n)` and `def Stirling3(n)`:

Comment on the accuracy of the approximation.

### Exercise 2. Round-off vs discretization error

Let  $f(x) = \sin(x)$  and  $x_0 = 1.2$ .

1. Write a Taylor expansion of  $f$  around  $x_0$  with a step size  $h$ , and derive a formula to approximate  $f'(x_0)$  of first order  $O(h)$  (see the lecture slides).
2. Plot the absolute error in a log-log scale as a function of the step size  $h$ , where  $h$  is in the range  $[10^{-16}, 1]$ . What is going on when  $h$  is too small?
3. Add another plot by changing the numpy floating point precision using `np.float32`, for both  $x_0$  and  $h$ . Please comment on your observations
  - use `np.logspace` to generate the  $h$  values

## Exercises on floating point systems

### Exercise 3. Decimal approximations

Compute decimal approximations of the real number 0.1 for 32-bit floating number precision in base 2. Then do the same thing for 0.25. Explain the difference that you observe. What about 0.35?

- You can use f-strings for printing the values `print(f'\nSingle Precision (32-bit): {single_precision:.20f}')`

### Exercise 4. Decimal Conversion of a Single-Precision Floating Point Number

Find the decimal equivalent of the following 32-bit precision machine number (sign, exponent, significand):

0 10000000 10010010000111111011011

- Mantissa 1's are in position 1,4,7,12-17,19-20, 22-23

This decimal number approximates a well-known number. How many significant digits does this 32-bit representation achieve?

## Exercises on stability and conditioning

### Exercise 5. Polynomials

Evaluate and plot the polynomial function  $(1 - x)^6$  written in its current factored and developed forms. Look in particular at the behavior close to the root  $x = 1$ . You can use the interval  $I = [0.995, 1.005]$  equally spaced by 100 points. Explain your observations.

- One can use sympy to find expression for the developed form

### Exercise 6. Catastrophic cancellation

Recall the quadratic formula to find the root of  $ax^2 + bx + c$ . Test the usual formula for the coefficients.  $a = 1$ ,  $b = -(10^8 + 10^{-8})$ ,  $c = 1$ . Explain what is going on, and propose a method to compute both roots accurately.